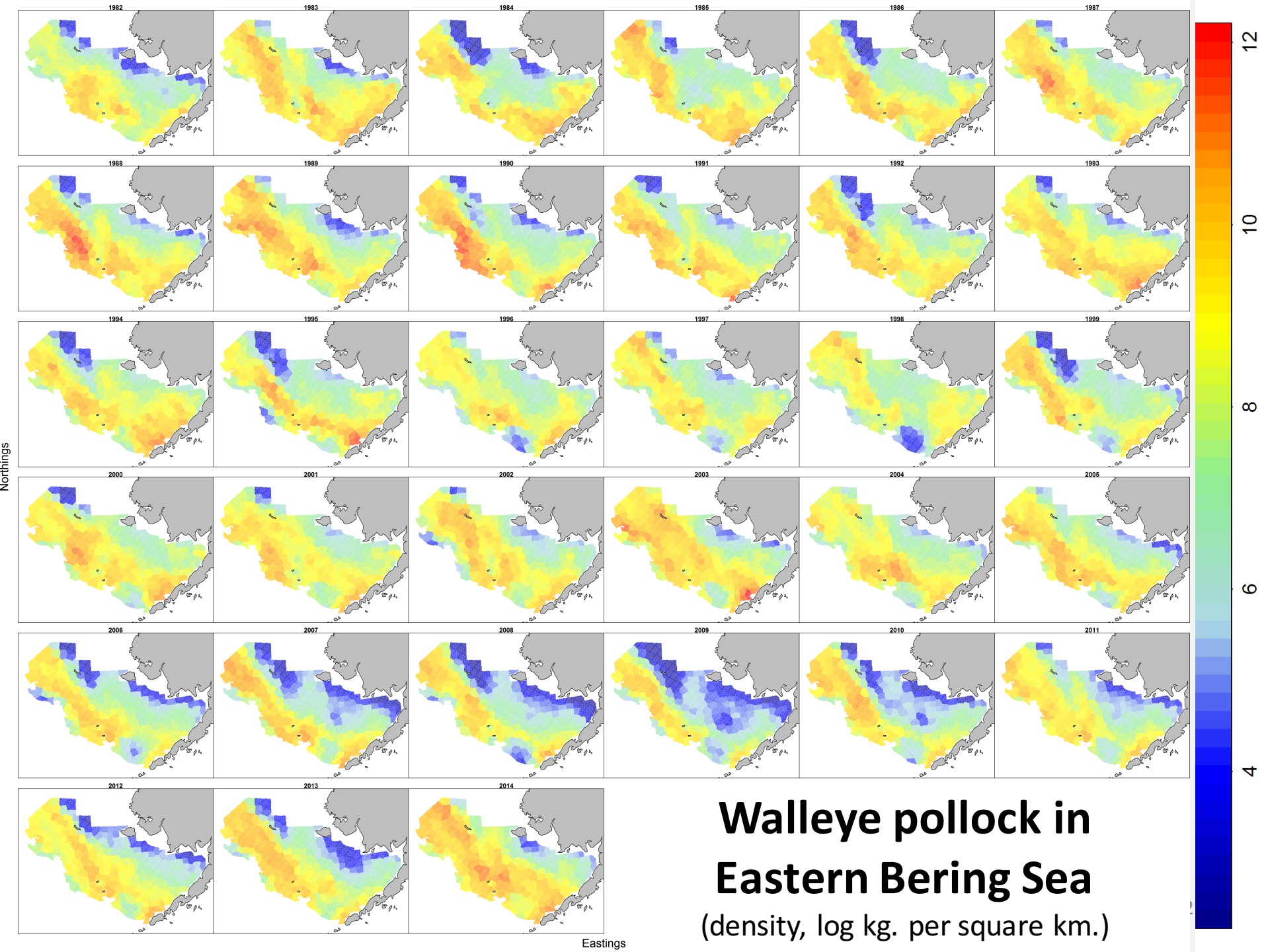


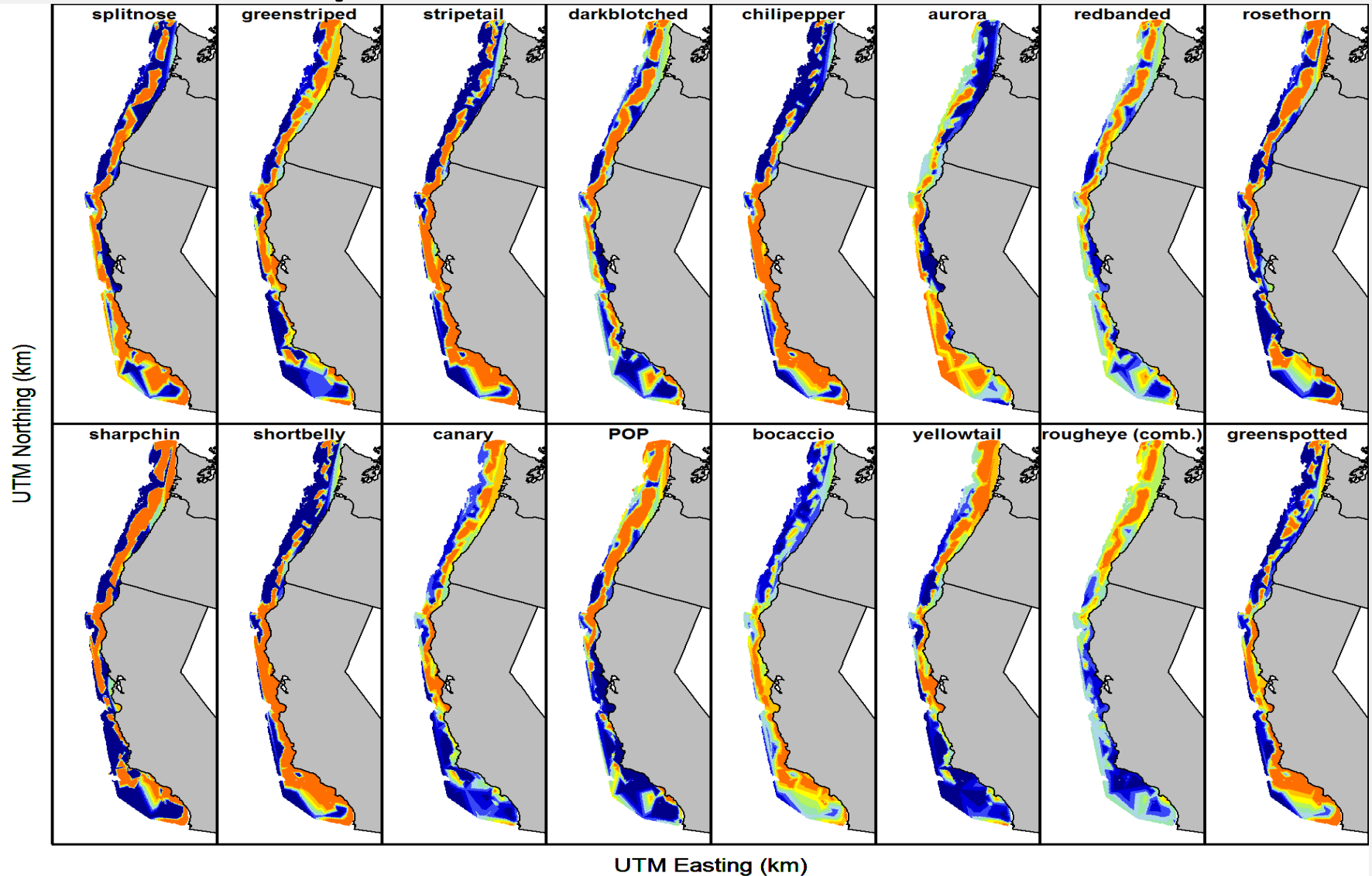
Developments in geostatistical modeling of survey data for pollock



Thorson, J.T., Ianelli, J.N., Larsen, E., Ries, L., Scheuerell, M.D., Szuwalski, C. & Zipkin, E. (In press) Joint dynamic species distribution models: a tool for community ordination and spatiotemporal monitoring. *Global Ecology and Biogeography*.



Joint species distribution models



Thorson, J.T., Scheuerell, M., Shelton, A.O., See, K., Skaug, H.J., Kristensen, K., In press. Spatial factor analysis: a new tool for estimating joint species distributions and correlations in species range. *Methods in Ecology and Evolution*

Joint species distribution models

Spatial factor analysis

- Estimates K “latent” maps for co-occurring species
- Separates process and measurement error
- Derived estimates of co-occurrence

Multispecies catches

$$c_p(s) \sim \text{Poisson}(\exp(\lambda_p(s)))$$

Expected densities

$$\lambda_p(s) = \alpha_p + \sum_{k=1}^K L_{p,k} \omega_k(s)$$

Where:

- $\omega_{i,k}$ is the k -th factor
- $L_{p,k}$ is the loadings for factor k on species p
- α_p is an intercept for species p
- $\lambda_{i,p}$ is predicted log-density for species p at location i

Spatial Gompertz model

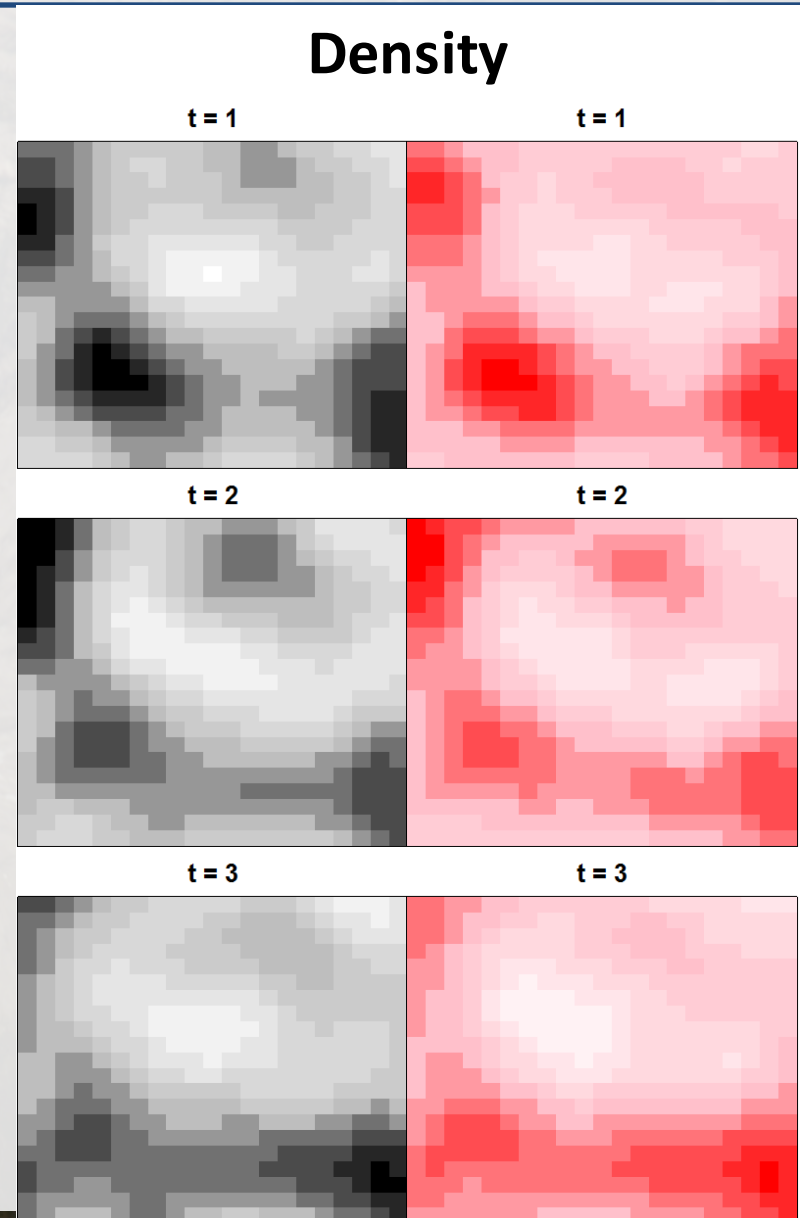
Autoregressive model

$$\log(\mathbf{d}_{t+1}) = \rho \log(\mathbf{d}_t) + \boldsymbol{\omega} + \boldsymbol{\varepsilon}_t$$

$$c_i \sim \text{Poisson} \left(d_{t(i)}(s_i) \right)$$

Where

- ρ is the strength of density dependence
- $\boldsymbol{\omega} \sim \text{GMRF}(\boldsymbol{\alpha}, \boldsymbol{\Sigma}_{\boldsymbol{\omega}})$ is spatial variation in carrying capacity
- $\boldsymbol{\varepsilon}_t \sim \text{GMRF}(\mathbf{0}, \boldsymbol{\Sigma}_{\boldsymbol{\varepsilon}})$ is spatiotemporal variation



Joint species distribution models

Model structure

- Factors follow Gompertz dynamics

$$\log(\mathbf{d}_{t+1,k}) = \rho \log(\mathbf{d}_{t,k}) + \boldsymbol{\omega}_k + \boldsymbol{\varepsilon}_{t,k}$$

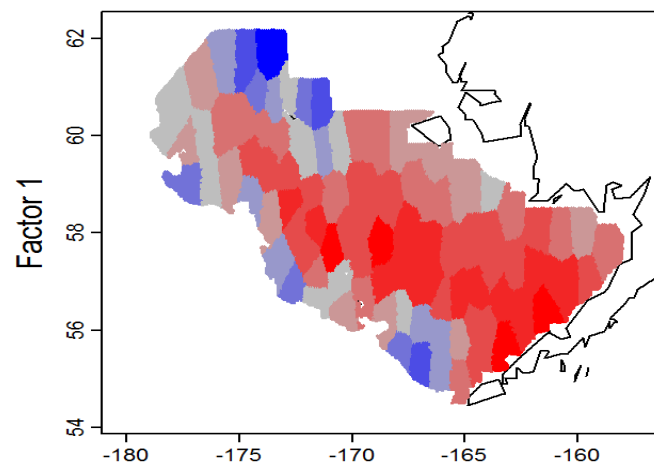
- Density is log-linear combination of factors

$$\lambda_{t,p}(s) = \alpha_p + \sum_{k=1}^{n_k} L_{p,k} \log(d_{t,k}(s))$$

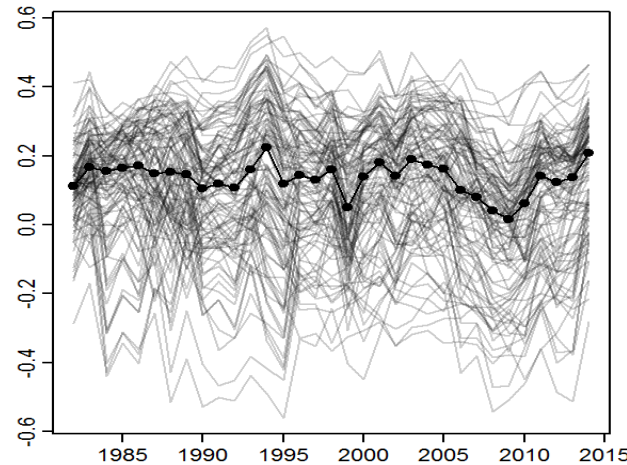
- Data follow a standard count process

$$c_{i,p} \sim \text{Poisson} \left(\lambda_{t(i),p}(s_i) \right)$$

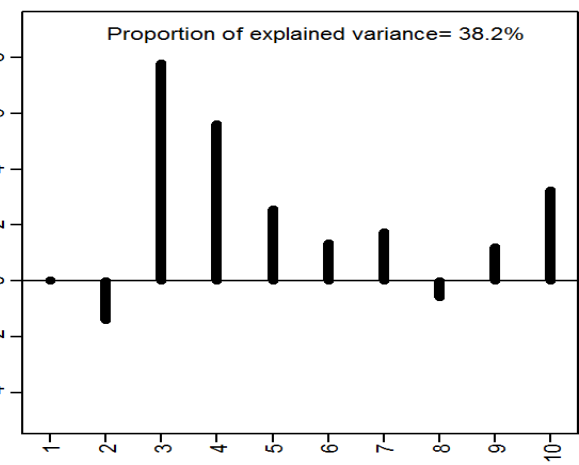
Average spatial effect



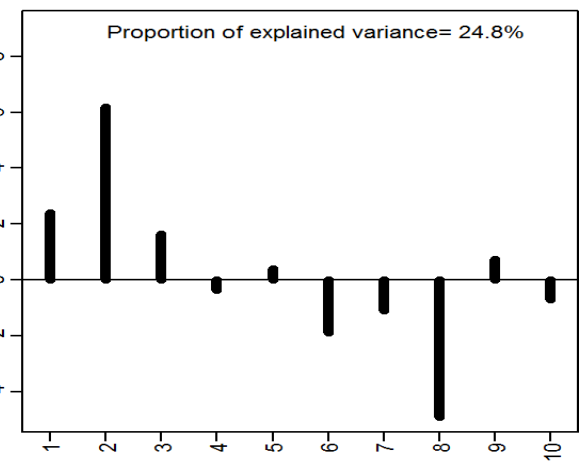
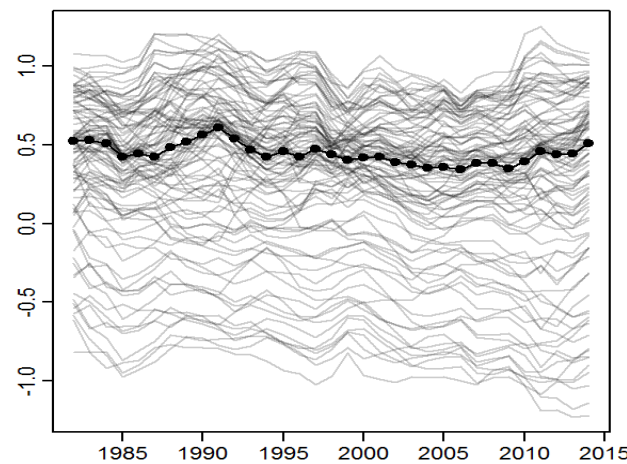
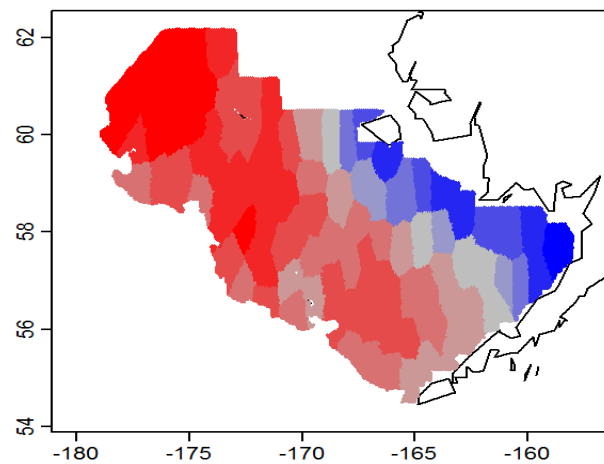
Average time trend



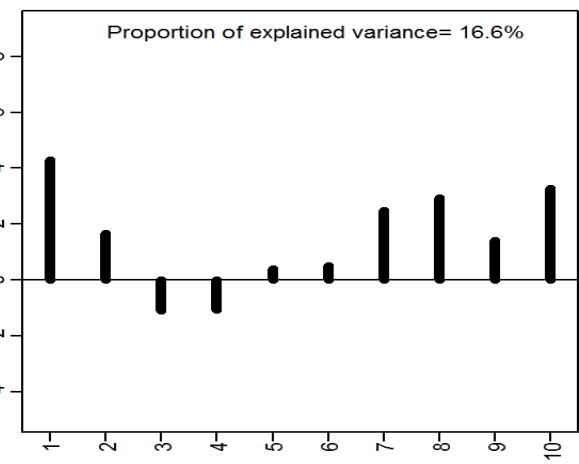
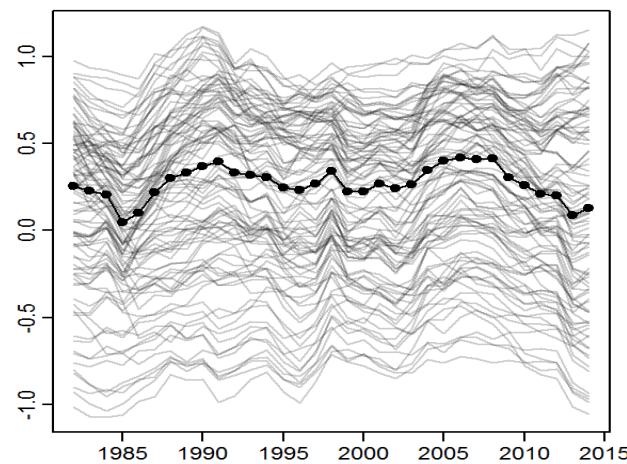
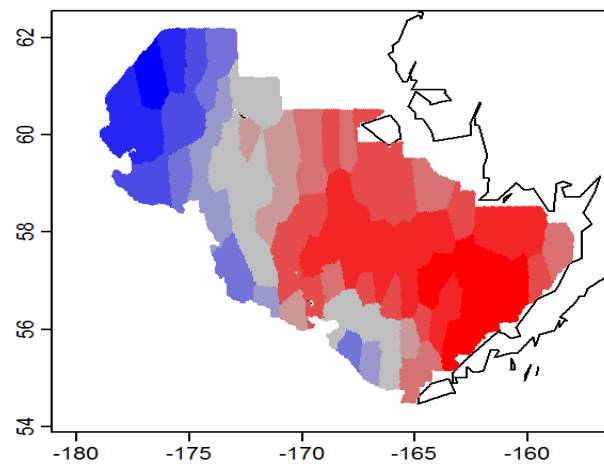
Factor loadings



Factor 2



Factor 3



Joint species distribution models

Conclusions

- Can reconstruct community dynamics
- Phylogenetic signal in species correlations
- Can identify “spatial scale” of community dynamics

		Species number									
Common Name		1	2	3	4	5	6	7	8	9	10
Tanner crab	1	1.000	0.454	0.049	-0.166	0.218	-0.176	0.047	0.067	-0.108	0.364
snow crab	2		1.000	-0.086	-0.277	-0.035	-0.411	-0.148	-0.638	0.356	-0.065
walleye pollock	3			1.000	0.910	0.505	0.130	0.227	-0.333	0.150	0.504
Pacific cod	4				1.000	0.592	0.502	0.196	-0.129	0.244	0.523
flathead sole	5					1.000	0.429	0.206	-0.068	0.271	0.153
Pacific halibut	6						1.000	0.041	0.522	0.251	0.441
yellowfin sole	7							1.000	0.346	0.649	0.510
starry flounder	8								1.000	-0.073	0.381
Alaska plaice	9									1.000	0.380
sturgeon poacher	10										1.000

Temperature impacts on pollock

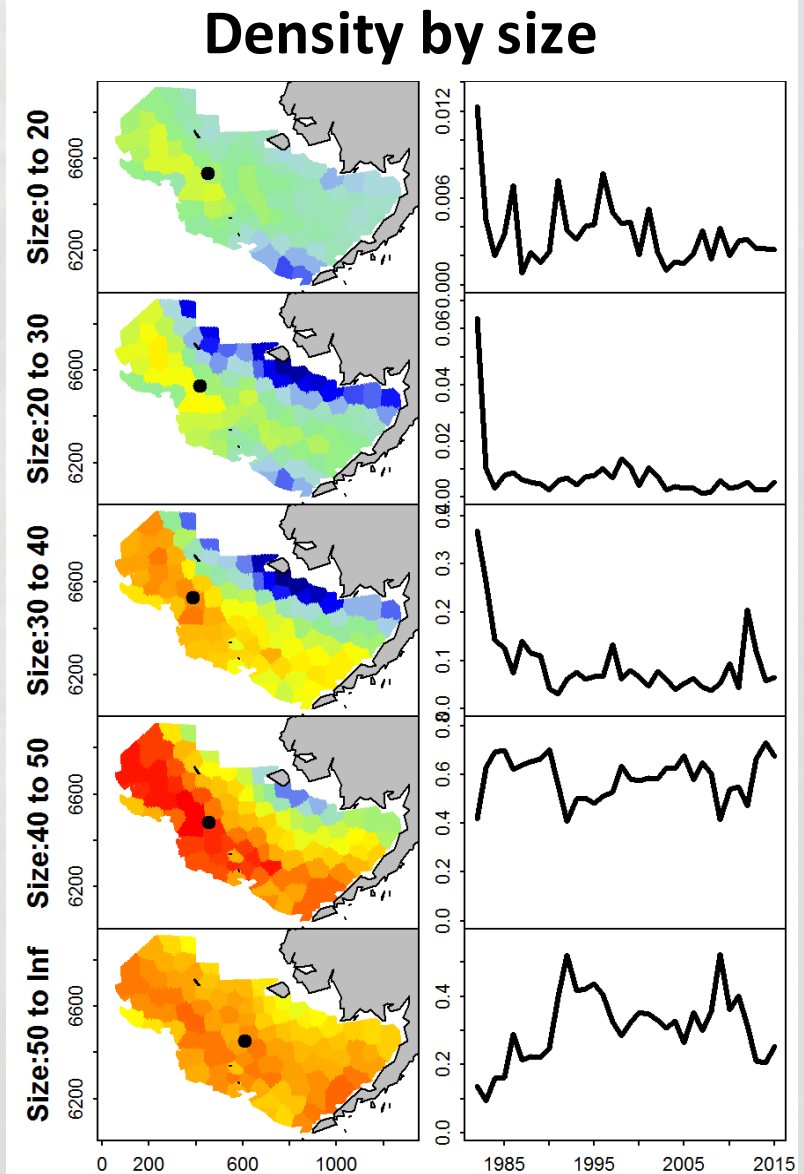


Thorson, J.T., Kotwicki, S, Ianelli, J.N. (In prep) The relative influence of temperature on fish distribution shifts: An important metric for forecasting future climate impacts.

Temperature impacts on pollock

VAST model

- Vector-Autoregressive Spatio-Temporal model
- Model density for every time, place, and size-class
 - Correlations over space for each size-bin
 - Correlations among size-bins for every year
- Includes temperature effects
 - Quadratic surface temperature
 - Quadratic bottom temperature
 - Regional temperature effect
 - Cold pool area interaction with location north-south or east-west



Temperature impacts on pollock

Derived quantities

$$d(s, t) = \exp(\lambda(s, t))$$

1. Total abundance

$$B(t) = \sum_{s=1}^{n_s} d(s, t) \times a(s)$$

2. Center of gravity

$$\bar{x}_t = \frac{1}{\sum_{s=1}^{n_s} (d(s, t) \times a(s))} \sum_{s=1}^{n_s} x(s) (d(s, t) \times a(s))$$

3. Average density

$$\bar{d}(t) = \frac{1}{\sum_{s=1}^{n_s} (d(s, t) \times a(s))} \sum_{s=1}^{n_s} d(s, t) (d(s, t) \times a(s))$$

Temperature impacts on pollock

Three counter-factuals

1. Just temperature effects

- Turn off spatio-temporal variation
- Turn off variation in size-structure over time

2. Just “recruitment-variation” effects

- Turn off temperature effects
- Turn off spatio-temporal variation

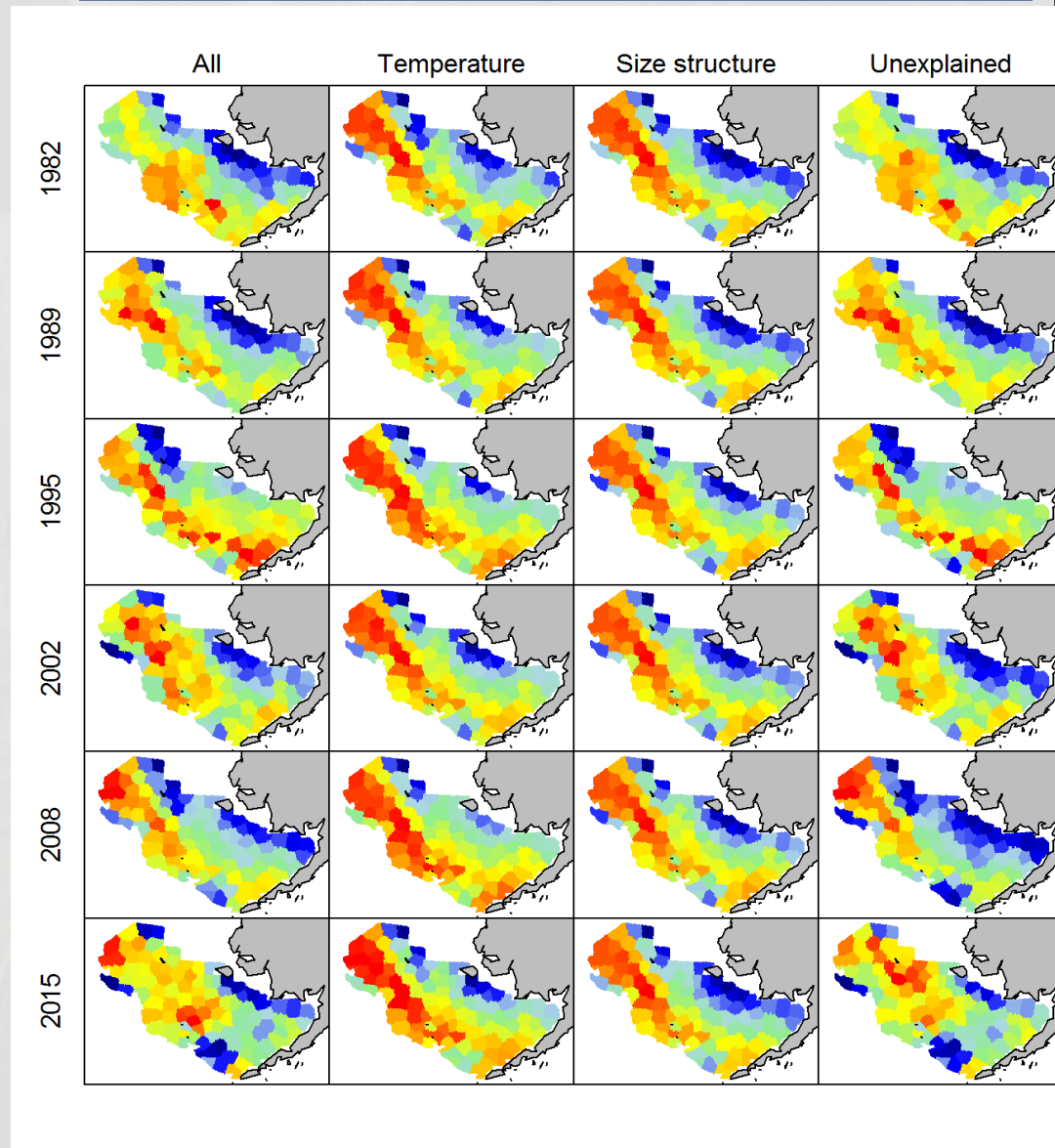
3. Just residual variation

- Turn off temperature effects
- Turn off variation in size-structure over time

Temperature impacts on pollock

Distribution for all counter-factuals

- Most observed variation is in the “unexplained” counter-factual



Temperature impacts on pollock

Distribution for all counter-factuals

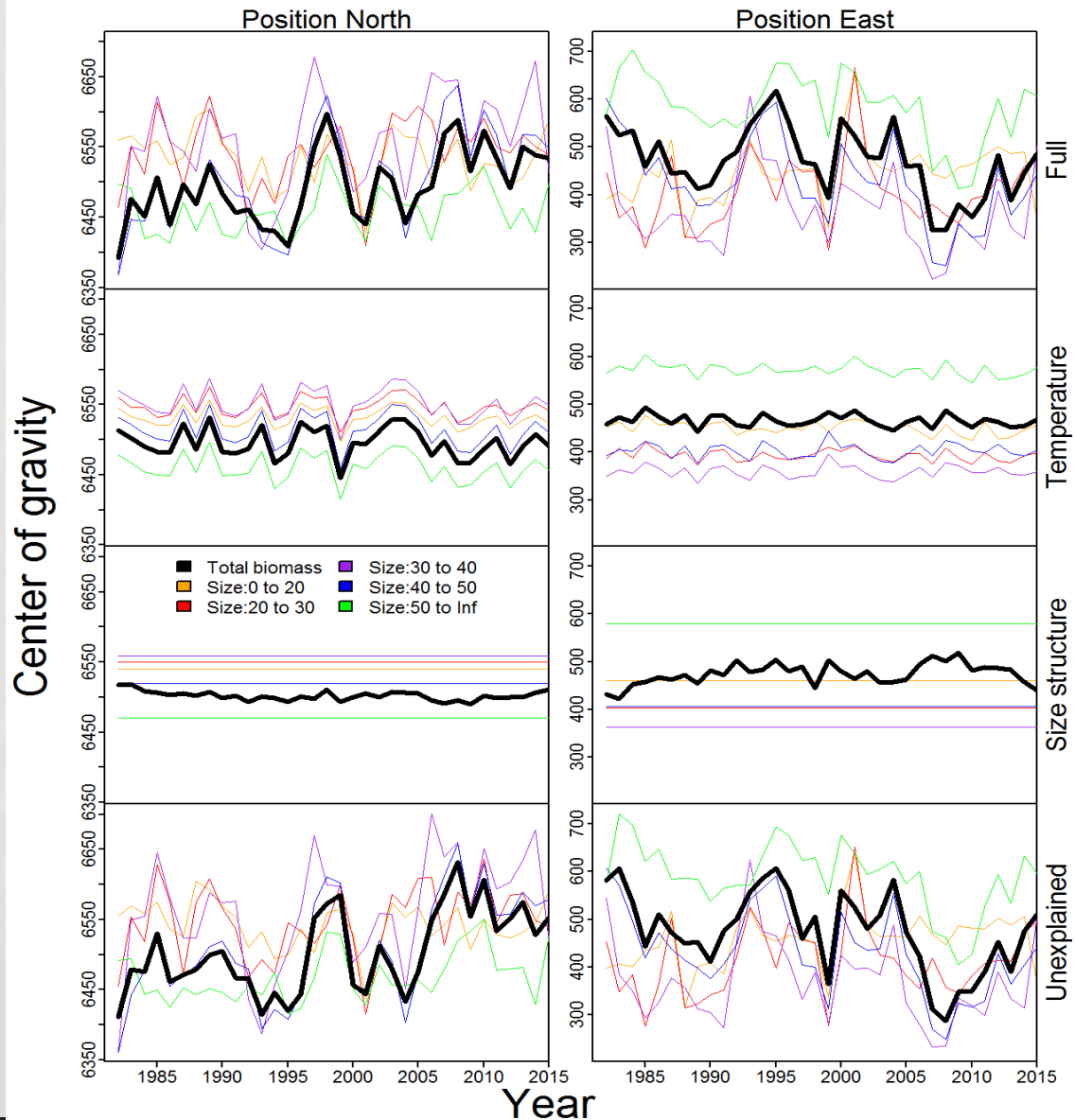
- Temperature generates some north-south variation
- Recruitment generates some east-west variation

	Northings COG (kilometres)	Eastings COG (kilometres)
Abundance weighted average estimator	44.6	70.4
Full model	51.5	74.0
Only temperature	21.5	12.8
Only recruitment-variation	6.8	22.7
Only unexplained	58.9	84.0

Temperature impacts on pollock

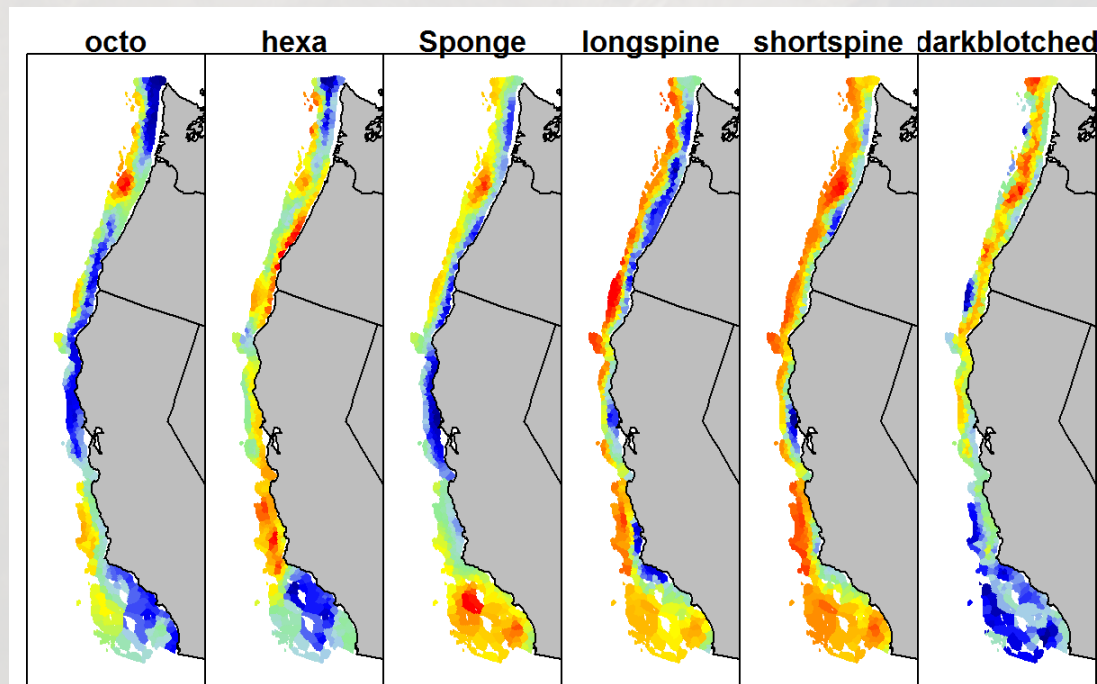
Distribution over time

- Trend north and west over time
- Neither temperature of size-structure captures the trend



Future directions

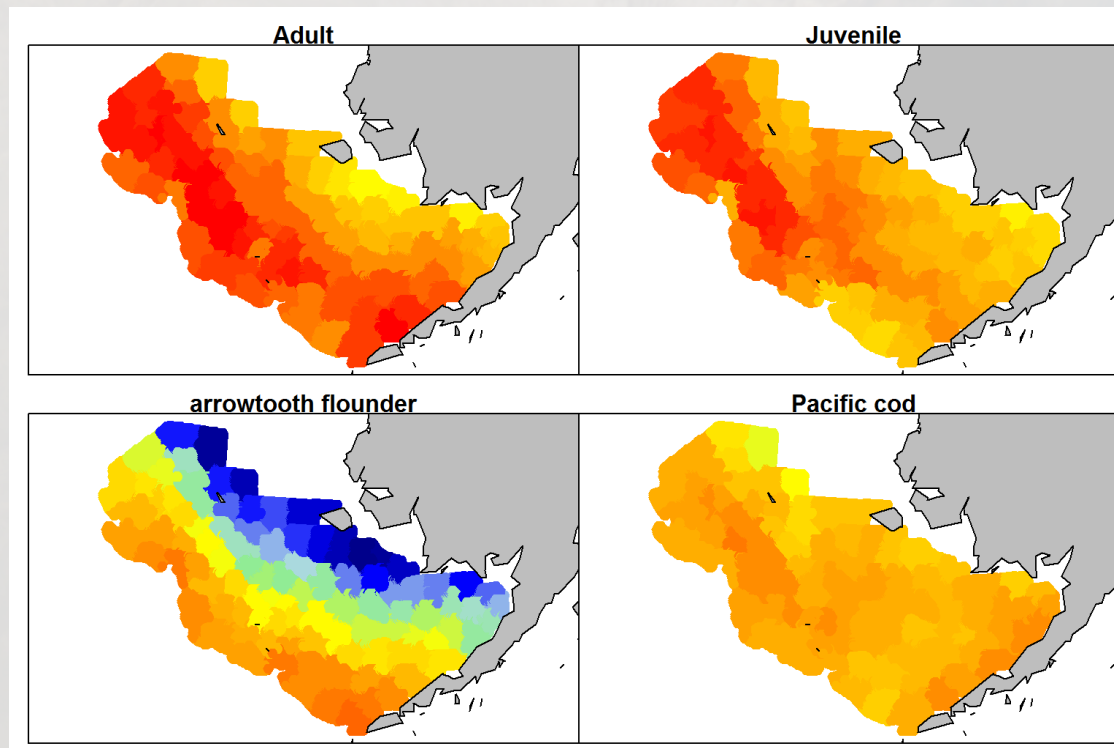
1. Spatial ecosystem-based management
 - Estimate associations between fish and habitat



Future directions

2. Spatial predator-prey overlap

- Estimate associations between fish and habitat



Acknowledgements

SDFA: Jim Ianelli, Mark Scheuerell, Elise Zipkin,
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